

EIGENVALUE SPACING DISTRIBUTION FOR THE ENSEMBLE OF REAL SYMMETRIC TOEPLITZ MATRICES

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ABSTRACT. Consider the ensemble of Real Symmetric Toeplitz Matrices, each entry iidrv from a fixed probability distribution p of mean 0, variance 1, and finite higher moments. The limiting spectral measure (the density of normalized eigenvalues) converges weakly to a new universal distribution with unbounded support, independent of p . This distribution's moments are almost those of the Gaussian's; the deficit may be interpreted in terms of Diophantine obstructions. With a little more work, we obtain almost sure convergence. An investigation of spacings between adjacent normalized eigenvalues looks Poissonian, and not GOE.

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1. INTRODUCTION

One of the central problems in Random Matrix Theory is as follows: consider some ensemble of matrices A with probabilities $p(A)$. As $N \rightarrow \infty$, what can one say about the density of normalized eigenvalues? For Real Symmetric matrices, where the entries are iidrv from suitably restricted probability distributions, the limiting distribution is the semi-circle. Note this ensemble has $\frac{N(N+1)}{2}$ independent parameters $(a_{ij}, i \leq j)$. For matrix ensembles with fewer degrees of freedom, different limiting distributions arise (for example, McKay [McK] proved d -regular graphs are given by Kesten's Measure). By examining ensembles with fewer than N^2 degrees

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of freedom, one has the exciting potential of seeing new, universal distributions. In this paper we investigate Symmetric Toeplitz matrices.

Definition 1.1. *A Toeplitz matrix is a matrix of the form*

$$\begin{pmatrix} b_0 & b_1 & b_2 & \cdots & b_{N-1} \\ b_{-1} & b_0 & b_1 & \cdots & b_{N-2} \\ b_{-2} & b_{-1} & b_0 & \cdots & b_{N-3} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ b_{1-N} & b_{2-N} & b_{3-N} & \cdots & b_0 \end{pmatrix} \quad (1)$$

We investigate symmetric Toeplitz matrices whose entries are chosen according to some distribution p with mean 0, variance 1, and finite higher moments. The probability density of a given matrix is $\prod_{i=0}^{N-1} p(b_i)$.

By looking at $\text{Trace}(A^2) = \sum_i \lambda_i^2(A)$, we see that the eigenvalues of A are of order \sqrt{N} . As the main diagonal is constant, all b_0 does is shift each eigenvalue. Therefore, it is sufficient to consider the case where the main diagonal vanishes.

To each Toeplitz matrix, we may attach a spacing measure by placing a point mass of size $\frac{1}{N}$ at each normalized eigenvalue:

$$\mu_{A,N}(x)dx = \frac{1}{N} \sum_{i=1}^N \delta\left(x - \frac{\lambda_i(A)}{\sqrt{N}}\right) dx. \quad (2)$$

The k^{th} moment of $\mu_{A,N}(x)$ is

$$M_k(A, N) = \frac{1}{N^{\frac{k}{2}+1}} \sum_{i=1}^N \lambda_i^k(A). \quad (3)$$

Let $M_k(N)$ be the average of $M_k(A, N)$ over the ensemble, with each A weighted by its density. We show that $M_k(N)$ converges to the moments of a new universal distribution, independent of p . The new distribution looks Gaussian, and numerical simulations and heuristic sketches at first seemed to support such a conjecture. A more detailed analysis, however, reveals that while $M_k(N)$ agrees with the Gaussian moments for odd k and $k = 0, 2$, the other even moments are less than the Gaussian.

We now sketch the proof. By the Trace Lemma,

$$\sum_{i=1}^N \lambda_i^k(A) = \text{Trace}(A^k) = \sum_{1 \leq i_1, \dots, i_k \leq N} a_{i_1, i_2} a_{i_2, i_3} \cdots a_{i_k, i_1}. \quad (4)$$

As our Toeplitz matrices are constant along diagonals, depending only on $|i_m - i_n|$, we have

$$M_k(N) = \frac{1}{N^{\frac{k}{2}+1}} \sum_{1 \leq i_1, \dots, i_k \leq N} \mathbb{E}(b_{|i_1-i_2|} b_{|i_2-i_3|} \cdots b_{|i_k-i_1|}), \quad (5)$$

where by $\mathbb{E}(\cdots)$ we mean averaging over the Toeplitz ensemble, with each matrix A weighted by its probability of occurring, and the b_j are iidrv drawn from $p(x)$.

We then show that as $N \rightarrow \infty$, the above sums vanish for k odd, and converge independent of p for k even to numbers M_k bounded by the moments of the Gaussian. By showing $\mathbb{E}[|M_k(A, N) - M_k(N)|^m]$ is small for $m = 2$ ($m = 4$), we obtain weak (almost sure) convergence.

Remark 1.2. *This problem was first posed by Bai [Bai], where he also asked similar questions about Hankel and Markov matrices. Almost surely the methods of this paper would be applicable to these cases. Bose and Bryc-Dembo-Jiang have independently observed that the limiting distribution is not Gaussian. Using a more probabilistic formulation, [BDJ] have calculated the moments using uniform variables and interpreting results as volumes of solids related to Eulerian numbers. We have independently found the same numbers, but through Diophantine analysis, which allows us to interpret the deviations from the Gaussian in terms of Diophantine obstructions, and estimate the rate of convergence.*

2. DETERMINATION OF THE MOMENTS

2.1. $k = 0, 2$ **and** k **odd.** For all N , $M_0(A, N) = M_0(N) = 1$. For $k = 2$, we have

$$M_2(N) = \frac{1}{N^2} \sum_{1 \leq i_1, i_2 \leq N} \mathbb{E}(b_{|i_1-i_2|} b_{|i_2-i_1|}) = \frac{1}{N^2} \sum_{1 \leq i_1, i_2 \leq N} \mathbb{E}(b_{|i_1-i_2|}^2). \quad (6)$$

As we have drawn the b s from a variance one distribution, the expected value above is 1 if $i_1 \neq i_2$ and 0 otherwise. Thus, $M_2(N) = \frac{N^2 - N}{N^2} = 1 - \frac{1}{N}$. Note there are two degrees of freedom. We can choose $b_{|i_1-i_2|}$ to be on any diagonal. Once we have specified the diagonal, we can then choose i_1 freely, which now determines i_2 .

For k odd, we must have at least one b_j occurring to an odd power. If one occurs to the first power, as the expected value of a product of independent variables is the product of the expected values, these terms contribute zero. Thus, the only contributions to an odd moment come when each b_j in the expansion occurs at least twice, and at least one occurs three times. Hence, if $k = 2m + 1$, we see we have at most $m + 1$ degrees of freedom, this

coming from the case $b_{j_1}^3 b_{j_2}^2 \cdots b_{j_m}^2$. There are m different factors of b , and then we can choose any one subscript. Once we have specified a subscript and which diagonals we are on, the remaining subscripts are determined. As all moments are finite, we find

$$M_{2m+1}(N) \ll_m \frac{1}{N^{\frac{2m+1}{2}+1}} N^{m+1} \ll_m \frac{1}{\sqrt{N}}. \quad (7)$$

2.2. Bounds for the Even Moments. We proceed in stages in calculating $M_{2k}(N)$, $2k \geq 4$. First, we bound $M_{2k}(N)$ by $2^k \cdot 2^k \cdot (2k-1)!!$, where $(2k-1)!!$ is the $2k^{\text{th}}$ moment of the Gaussian. We then show that each factor of 2^k can be removed, and then show a strict inequality holds.

$$M_{2k}(N) = \frac{1}{N^{k+1}} \sum_{1 \leq i_1, \dots, i_{2k} \leq N} \mathbb{E}(b_{|i_1-i_2|} b_{|i_2-i_3|} \cdots b_{|i_{2k}-i_1|}). \quad (8)$$

If any b_j occurs to the first power, its expected value is zero and there is no contribution. Thus, the b_j s must be matched at least in pairs. If any b_j occurs to the third or higher power, there are less than $k+1$ degrees of freedom, and there will be no contribution in the limit.

The b_j s are matched in pairs, say $b_{|i_m-i_{m+1}|} = b_{|i_n-i_{n+1}|}$. Let $x_m = |i_m - i_{m+1}| = |i_n - i_{n+1}|$. There are two possibilities:

$$i_m - i_{m+1} = i_n - i_{n+1} \quad \text{or} \quad i_m - i_{m+1} = -(i_n - i_{n+1}). \quad (9)$$

There are k such pairs, thus we have 2^k choices of sign. Further, there are $(2k-1)!!$ ways to pair off $2k$ numbers into groups of two.

Fix a choice of sign and a pairing. Once we specify x_1, \dots, x_k and any one index, say i_1 , all the other indices are almost determined (if the choices are consistent). There is one remaining freedom. After we've chosen which differences to match and the values of these differences and the choice of signs, for each time when there is a negative sign, there is one additional choice: does the positive or negative difference occur first? Thus, after we specify for each pair whether the positive or negative difference occurs first, then all the indices are determined.

Therefore, there are N^{k+1} degrees of freedom. If all the x_j s are distinct, we have the expected value of the second moment of p , k times. These contribute at most

$$\frac{1}{N^{k+1}} \cdot 2^k \cdot 2^k (2k-1)!! N^{k+1} = 2^k (2k-1)!!. \quad (10)$$

If some of the x_j s are equal, we have fewer than $k+1$ degrees of freedom. We now have the expected value of a product of moments of p , which is

finite and independent of N . These terms will not contribute in the limit. Therefore

$$\lim_{N \rightarrow \infty} M_{2k}(N) \leq 2^k \cdot 2^k (2k-1)!!. \quad (11)$$

We now remove the factor of 2^k coming from the choice of signs. Consider a pairing of the b_j s. We claim the only term which contributes in the limit is when all signs are negative.

Let x_1, \dots, x_k be the values of the $|i_j - i_{j+1}|$ s, and let $\epsilon_1, \dots, \epsilon_k$ be the choices of sign (see Equation 9). Define $\tilde{x}_1 = i_1 - i_2, \tilde{x}_2 = i_2 - i_3, \dots, \tilde{x}_{2k} = i_{2k} - i_1$. Note exactly one \tilde{x}_j is x_j and exactly one is $\epsilon_j x_j$. We have

$$\begin{aligned} i_2 &= i_1 - \tilde{x}_1 \\ i_3 &= i_1 - \tilde{x}_1 - \tilde{x}_2 \\ &\vdots \\ i_1 &= i_1 - \tilde{x}_1 - \dots - \tilde{x}_{2k}. \end{aligned} \quad (12)$$

Therefore

$$\tilde{x}_1 + \dots + \tilde{x}_{2k} = \sum_{j=1}^k (1 + \epsilon_j) x_j = 0. \quad (13)$$

If any $\epsilon_j = 1$, then the x_j are not linearly independent, and we have fewer than $k+1$ degrees of freedom; these terms will not contribute in the limit. Thus, the only valid assignment is to have all the signs negative. There are now 2^k possible choices of order (whether the negative or positive difference occurs first), giving $2^k \cdot N^{k+1}$. We eliminate 2^k by changing our viewpoint.

We have $k+1$ degrees of freedom. We match our differences into k pairs. Choose i_1 and i_2 . We now look at the freedom to choose the remaining indices i_j . Once i_1 and i_2 are specified, we have $i_1 - i_2$, and a later difference must be the negative of that. If $i_2 - i_3$ is matched with $i_1 - i_2$, then i_3 is uniquely determined (because it must give the opposite of the earlier difference). If not, i_3 is a new variable. Now look at i_4 . If $i_3 - i_4$ is matched with an earlier difference, then the sign of its difference is known, and i_4 is uniquely determined; if this difference belongs to a new pair not previously encountered, then i_4 is a new variable and free. Proceeding in this way, we note that if we encounter i_n such that $i_{n-1} - i_n$ is paired with a previous difference, the sign of its difference is specified, and i_n is uniquely determined; otherwise, if this is a difference of a new pair, i_n is a free variable, with at most N choices. Thus we see there are at most N^{k+1} choices (note not all

choices will work, as for example the final difference $i_{2n} - i_1$ is determined before we get there, because of earlier choices).

More explicitly, having $k + 1$ degrees of freedom does not imply each term contributes fully – we will see there are Diophantine obstructions which bound the moments away from the Gaussian's. However, each pairing and choice of sign contributes at most N^{k+1} , and we have shown

$$M_{2k}(N) \leq (2k - 1)!! + O_k\left(\frac{1}{N}\right). \quad (14)$$

2.3. The Fourth Moment. The fourth moment calculation highlights the Diophantine obstructions encountered, which bound the moments away from the Gaussian.

$$M_4(N) = \frac{1}{N^3} \sum_{1 \leq i_1, i_2, i_3, i_4 \leq N} \mathbb{E}(b_{|i_1 - i_2|} b_{|i_2 - i_3|} b_{|i_3 - i_4|} b_{|i_4 - i_1|}) \quad (15)$$

Let $x_j = |i_j - i_{j+1}|$. If any b_{x_j} occurs to the first power, its expected value is zero. Thus, either the x_j are matched in pairs (with different values), or all four are equal (in which case they are still matched in pairs). There are 3 possible matchings; however, by symmetry (simply relabel), we see the contribution from $x_1 = x_2, x_3 = x_4$ is the same as the contribution from $x_1 = x_4, x_2 = x_3$.

If $x_1 = x_2, x_3 = x_4$, we have

$$i_1 - i_2 = -(i_2 - i_3) \quad \text{and} \quad i_3 - i_4 = -(i_4 - i_1). \quad (16)$$

Thus, $i_1 = i_3$ and i_2 and i_4 are arbitrary. Using these three variables as our independent degrees of freedom, we see there are N^3 such quadruples. Almost all of these will have $x_1 \neq x_3$, and contribute $\mathbb{E}(b_{x_1}^2 b_{x_3}^2) = 1$. Given i_1 and i_2 , $N - 1$ choices of i_4 yield $x_1 \neq x_3$, and one choice yields the two equal. Letting p_4 denote the fourth moment of p , we see this case contributes

$$\frac{1}{N^3} \left(N^2(N - 1) \cdot 1 + N^2 \cdot p_4 \right) = 1 - \frac{1}{N} + \frac{p_4}{N} = 1 + O\left(\frac{1}{N}\right). \quad (17)$$

The other possibility is for $x_1 = x_3$ and $x_2 = x_4$. Non-adjacent pairing is what leads to Diophantine obstructions, which decreases the contribution to the moment. Now we have

$$i_1 - i_2 = -(i_3 - i_4) \quad \text{and} \quad i_2 - i_3 = -(i_4 - i_1). \quad (18)$$

This yields

$$i_1 = i_2 + i_4 - i_3, \quad i_1, i_2, i_3, i_4 \in \{1, \dots, N\}. \quad (19)$$

The fact that each $i_j \in \{1, \dots, N\}$ is what leads to the Diophantine obstructions. In the first case, we saw we had three independent variables, and $N^3 + O(N^2)$ choices that were mutually consistent. Now, it is possible for choices of i_2, i_3 and i_4 to lead to impossible values for i_1 . For example, if $i_2, i_4 \geq \frac{2N}{3}$ and $i_3 < \frac{N}{3}$, we see $i_1 > N$. Thus, there are at most $(1 - \frac{1}{27})N^3$ valid choices. This is enough to show the Gaussian moment is strictly greater; later we will see that if there is one moment less than the Gaussian, all larger even moments are also smaller.

The following lemma shows this case contributes $\frac{2}{3}$ to the fourth moment.

Lemma 2.1. *Let $I_N = \{1, \dots, N\}$. Then $\#\{x, y, z \in I_N : 1 \leq x + y - z \leq N\} = \frac{2}{3}N^3 + \frac{1}{3}N$.*

Proof. Say $x + y = S \in \{2, \dots, 2N\}$. For $2 \leq S \leq N$, there are $S - 1$ choices of z , and for $S \geq N + 1$, there are $2N - S + 1$. Similarly, the number of $x, y \in I_N$ with $x + y = S$ is $S - 1$ if $S \leq N + 1$ and $2N - S + 1$ otherwise. The number of triples is

$$\sum_{S=2}^N (S - 1)^2 + \sum_{S=N+1}^{2N} (2N - S + 1)^2 = \frac{2}{3}N^3 + \frac{1}{3}N. \quad (20)$$

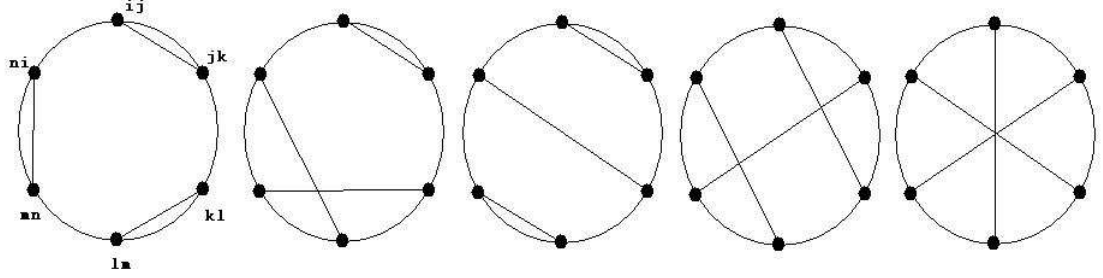
□

Collecting all the pieces, we have shown

Theorem 2.2 (Fourth Moment). *Let p_4 be the fourth moment of p . Then*

$$M_4(N) = 2\frac{2}{3} + \frac{2(p_4 - 1)}{N} + \frac{1}{N^2}. \quad (21)$$

2.4. Sixth and Eighth Moments. Any even moment can be explicitly determined by brute-force calculation, though deriving exact formulas as $k \rightarrow \infty$ requires handling involved combinatorics. To calculate the higher moments, consider $2k$ points on the unit circle, and look at how many different shapes we get when we match in pairs. We find $M_6(N) = 11$ (compared to the Gaussian's 15), and $M_8(N) = 64\frac{4}{15}$ (compared to the Gaussian's 105). For the sixth moment, there are five different configurations:



These occur 2, 6, 3, 3 and 1 time, contributing $1, \frac{2}{3}, 1, \frac{1}{2}$, and $\frac{1}{2}$ (respectively); these correspond to the $15 = (6 - 1)!!$ pairings. For the eight moment, the smallest contribution is $\frac{1}{4}$, coming from the matching $x_1 = x_3, x_2 = x_4, x_5 = x_7, x_6 = x_8$. It seems the more crossings (in some sense), the greater the Diophantine obstructions and the smaller the contribution.

3. UPPER BOUNDS OF HIGH MOMENTS

3.1. Weak Upper Bound of High Moments.

Lemma 3.1. *For $2k \geq 4$, $\lim_{N \rightarrow \infty} M_{2k}(N) < (2k - 1)!!$.*

Proof. Once we find a pairing that contributes less than 1 for some moment, we note that it will lift to pairings for higher moments that will also contribute less than 1. Say we have such a pairing on $b_{|i_1 - i_2|} \cdots b_{|i_{2k_0} - i_1|}$ giving less than 1. We extend this to a pairing on $2k > 2k_0$ as follows. We now have

$$b_{|i_1 - i_2|} \cdots b_{|i_{2k_0-1} - i_{2k_0}|} b_{|i_{2k_0} - i_{2k_0+1}|} b_{|i_{2k_0+1} - i_{2k_0+2}|} \cdots b_{|i_{2k-1} - i_{2k}|} b_{|i_{2k} - i_i|}. \quad (22)$$

In groups of two, pair adjacent neighbors from $b_{|i_{2k_0+1} - i_{2k_0+2}|}$ to $b_{|i_{2k-1} - i_{2k}|}$. This implies $i_{2k_0} = i_{2k_0+2} = \cdots = i_{2k}$. Thus, looking at the first $2k_0 - 1$ and the last factor gives

$$b_{|i_1 - i_2|} \cdots b_{|i_{2k_0-1} - i_{2k_0}|} b_{|i_{2k} - i_i|} = b_{|i_1 - i_2|} \cdots b_{|i_{2k_0-1} - i_{2k_0}|} b_{|i_{2k_0} - i_i|}. \quad (23)$$

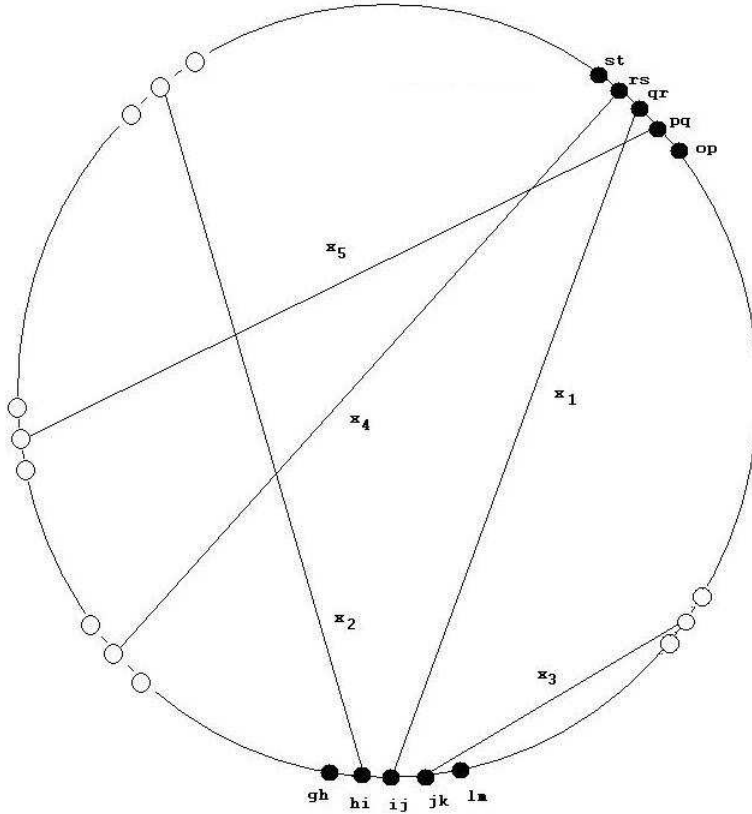
Now pair these as in the pairing which gave less than 1, and we see this pairing will contribute less than 1 as well. \square

3.2. Strong Upper Bound of High Moments. In general, the further away one moment is from the Gaussian, the more one can say about higher moments. While we do not have exact asymptotics, one can show

Theorem 3.2. $\lim_{n \rightarrow \infty} \frac{M_{2k}}{(2k-1)!!} = 0$.

Proof. We will show that for any positive integer c , for k sufficiently large, as $N \rightarrow \infty$ the moment is bounded by $(\frac{2}{3})^c (2k-1)!!$. We have shown that we may take as independent variables the k values of the subscripts of the b_j s (x_1, \dots, x_k) and any index. The goal is to show that almost all of the pairings, for k large, have at least c Diophantine obstructions (of the type encountered in the fourth moment). If there were no obstructions, these terms would contribute N^3 ; the obstructions reduce the contribution to $\frac{2}{3}N^3$.

We strategically replace our set of independent variables i_d, x_1, \dots, x_k with new variables which exhibit the obstructions. We give full details on dealing with one obstruction, and sketch how to add more. For simplicity, instead of referring to i_1, i_2, \dots, i_{2k} , we use i, j, k, \dots and p, q, r, \dots . Thus, in the trace expansion we have terms like $a_{i_1 i_2} = b_{|i_1 - i_2|}$; we refer to this point by $i_1 i_2$ or by ij .



Say we pair $b_{|i-j|}$ with $b_{|q-r|}$. Let $x_1 = i - j = -(q - r)$. If we knew $i = j + r - q$, with j, r and q independent free variables, then our earlier results show there are only $\frac{2}{3}N^3$, not N^3 , solutions. Unfortunately, j, r and

q need not be independent; however, for almost all of the $(2k-1)!!$ pairings, they will be.

Create a buffer zone around ij and qr of two vertices on each side, and assume that neither buffer zone intersects. Given ij , there are $(2k-1)-8$ possible choices to place qr . Now connect the neighbors of ij and qr such that nothing is connected within one vertex of another. There will be $(2k-O(1)) \cdot (2k-O(1)) \cdot (2k-O(1)) \cdot (2k-O(1))$ such pairings. Note that, as we start placing some of these connections, some vertices become unavailable. For example, say there is exactly one vertex between the buffer of ij and the buffer of qr . This vertex is *not* available for use, for if we were to place another vertex there, the indices it gives would not be independent. The same would be true if there were just two vertices between the two buffers, and so on. In each case, however, we only lose $O(1)$ vertices. As all these pairings are separated, we may label their differences by x_2, x_3, x_4 and x_5 , independent free variables.

The point is that the separation allows us to replace some the independent variables x_d with j, r and q . Note that each index appears in exactly two vertices on the circle, and they are adjacent. Thus, these are the only occurrences of i, j, q, r and we may replace x_5 with q , x_4 with r , and x_1 with j . We now have the desired situation: $i = j + r - q$, with all three on the left independent free variables.

There are $(2k-11)!!$ ways to pair the remaining vertices. For those pairs that have j, q, r independent, the contribution is at most $\frac{2}{3}N^3 \cdot N^{k+1-3}$; for the others, we bound the contribution by N^{k+1} . Hence

$$\begin{aligned} M_{2k}(N) &\leq \frac{1}{N^{k+1}} \left[(2k)^5 (2k-11)!! \frac{2}{3} N^{k+1} + O(k^4) \cdot (2k-11)!! \cdot N^{k+1} \right] \\ &\leq \frac{2}{3} (2k-1)!! + O\left(\frac{(2k-1)!!}{k}\right). \end{aligned} \quad (24)$$

Therefore,

$$\frac{M_{2k}(N)}{(2k-1)!!} \leq \frac{2}{3} + O\left(\frac{1}{k}\right). \quad (25)$$

There are two ways to handle the general case with c Diophantine obstructions. One may start with enormous buffer zones around the initial pairs. As the construction progresses, we open up more and more portions of the parts of the buffer zones not immediately near the vertices. This keeps all but $O(1)$ vertices available for use. Alternatively, along the lines of the first construction, we can just note that by the end of stage c , $O_c(1)$ vertices were unusable. We will still have the correct power of $2k$, with a correction term smaller by a factor of $\frac{1}{k}$. \square

4. LOWER BOUND OF HIGH MOMENTS

4.1. Preliminaries. By obtaining a sufficiently large lower bound for the even moments, we show the limiting distribution has unbounded support. In particular, we must find a lower bound L_{2k} such that $\lim_{k \rightarrow \infty} \sqrt[2k]{L_{2k}} = \infty$.

We know the moments are bounded by those of the Gaussian, $(2k-1)!!$; the limiting value of the $2k$ -th root of the Gaussian (by Stirling's Formula) is $\frac{k}{e}$. We will show $\sqrt[2k]{L_{2k}} \approx k^{\frac{1}{2}-\epsilon}$ in the limit.

The construction is as follows: in studying the $2k$ -th moment, we are led to sums of the form

$$\begin{aligned} & \frac{1}{N^{k+1}} \mathbb{E} \left[\sum_{i_1=1}^N \cdots \sum_{i_{2k}=1}^N a_{i_1, i_2} a_{i_2, i_3} \cdots a_{i_{2k}, i_1} \right] \\ &= \frac{1}{N^{k+1}} \mathbb{E} \left[\sum_{i_1=1}^N \cdots \sum_{i_{2k}=1}^N b_{|i_1 - i_2|} b_{|i_2 - i_3|} \cdots b_{|i_{2k} - i_1|} \right]. \end{aligned} \quad (26)$$

If any $b_{|i_n - i_{n+1}|}$ occurs only once, as it is drawn from a mean zero distribution, there is no contribution to the expected value. Thus, the $2k$ numbers (the b s) are matched in at least pairs, and, to obtain a lower bound, it is sufficient to consider the case where the differences are matched in k pairs. Let these positive differences (of $|i_n - i_{n+1}|$) be x_1, \dots, x_k .

In Section 2.2, we showed the matchings must occur with negative signs. Thus, if $|i_n - i_{n+1}| = |i_y - i_{y+1}|$, then $(i_n - i_{n+1}) = -(i_y - i_{y+1})$. We let $\tilde{x}_j = i_j - i_{j+1}$. Thus, for any x_j , there is a unique j_1 such that $\tilde{x}_{j_1} = x_j$, and a unique j_2 such that $\tilde{x}_{j_2} = -x_j$. We call the first set of differences positive, and the other set negative; we often denote these by \tilde{x}_p and \tilde{x}_n , and note that we have k of each.

We have $k+1$ degrees of freedom. We may take these as the k differences x_k , and then any index, say i_1 . We have the relations

$$\begin{aligned} i_2 &= i_1 - \tilde{x}_1 \\ i_3 &= i_1 - \tilde{x}_1 - \tilde{x}_2 \\ &\vdots \\ i_{2k} &= i_1 - \tilde{x}_1 - \cdots - \tilde{x}_{2k}. \end{aligned} \quad (27)$$

Once we specify i_1 and the differences \tilde{x}_1 through \tilde{x}_{2k} , all the indices are determined. If everything is matched in pairs and each $i_j \in \{1, \dots, N\}$, then we have a valid configuration, which will contribute $+1$ to the $2k$ -th moment. The reason it contributes $+1$ is because, as everything is matched in pairs, we have the expected value of the second moment of $p(x)$, k times.

Thus, we need to show the number of valid configurations is sufficiently large. The problem is that, in Equation 27, each index $i_j \in \{1, \dots, N\}$; however, it is possible that a running sum $i_1 - \tilde{x}_1 - \dots - \tilde{x}_m$ is not in this range for some m . We will show that we are often able to keep all these running sums in the desired range.

4.2. Construction. Let $\alpha \in (\frac{1}{2}, 1)$. Let $I_A = \{1, \dots, A\}$, where $A = \frac{N}{k^\alpha}$. Choose each difference x_j from I_A ; there are A^k ways to do this. In the end, we want to study k -tuples such that no value is chosen twice. Note such tuples are lower order, namely there are at most $\binom{k}{2} A^{k-1}$ such tuples. This is $O(N^{k-1})$. As i_1 takes on at most N values (not all values will in general lead to valid configurations), we see tuples with repeated values occur at most $O(N^k)$ times; as we divide by N^{k+1} , these terms will not contribute for fixed k as $N \rightarrow \infty$. Thus, with probability one (as $N \rightarrow \infty$), we may assume the k values x_j are distinct.

Let us consider k distinct positive numbers (the x_j s) drawn from I_A , giving rise to k positive differences \tilde{x}_p s and k negative differences \tilde{x}_n s. Let us make half of the numbers $\tilde{x}_1, \dots, \tilde{x}_k$ positive (arising from the \tilde{x}_p s), and half of these numbers negative (arising from the \tilde{x}_n s). Call this the first block (of differences).

Then, in the differences $\tilde{x}_{k+1}, \dots, \tilde{x}_{2k}$ (the second block), we have the remaining differences. Note every positive (negative) difference in $\tilde{x}_1, \dots, \tilde{x}_k$ is paired with a negative (positive) difference in $\tilde{x}_{k+1}, \dots, \tilde{x}_{2k}$. Note we have not specified the order of the differences, just how many positive (negative) are in the first block / second block.

Note two different k -tuples of differences x_j *cannot* give rise to the same configuration (if we assume the differences are distinct). This trivially follows from the fact that the differences specify which diagonal of the Toeplitz matrix the $a_{i_m i_{m+1}}$ s are on; if we have different tuples, there is at least one diagonal with an entry on one but not on the other.

Let us assume we have chosen the order of the differences in the first block, $\tilde{x}_1, \dots, \tilde{x}_k$. We look at a subset of possible ways to match these with differences in the second block. In the second block, there are $\frac{k}{2}$ positive (negative) differences \tilde{x}_p (\tilde{x}_n). There are $(\frac{k}{2})!$ ways to choose the relative order of the positive (negative) differences. Note we are *not* giving a complete ordering of the differences in the second block. There are $k! > (\frac{k}{2})!^2$ ways to completely order. We are merely specifying the relative order among the positive (negative) elements, and not specifying how the positive and negative differences are interspersed.

Thus, the number of matchings, each of which contribute 1, obtainable by this method is at most

$$N \cdot (A^k - O(A^{k-1})) \cdot (k/2)!^2, \quad (28)$$

where N is from the possible values for i_1 , $A^k - O(A^{k-1})$ is the number of k -tuples of distinct differences $x_j \in I_A$, and $(k/2)!^2$ is the number of relative arrangements of the positive and negative differences in the second block (each of which is matched with an opposite difference in the first block).

Not all of the above will yield a $+1$ contribution to the $2k$ -th moment. Remember, each index i_m must be in $\{1, \dots, N\}$. We now show that for a large number of the above configurations, we do have all indices appropriately restricted. We call such a configuration *valid*.

4.3. Number of Valid Configurations. Most of the time, the sum of the positive differences \tilde{x}_p in the first block will be close to the negative of the sum of the negative differences \tilde{x}_n in the first block.

Explicitly, we may regard the \tilde{x}_p s (\tilde{x}_n s) as independent random variables taken from the uniform distribution on I_A ($-I_A$) with mean approximately $\frac{1}{2}A$ ($-\frac{1}{2}A$) and standard deviation approximately $\frac{1}{2\sqrt{3}}A$. By the Central Limit Theorem, for k large, the sum of the $\frac{k}{2}$ positive (negative) \tilde{x}_p s (\tilde{x}_n s) in the first block converges to a normal distribution with mean approximately $\frac{kA}{4}$ ($-\frac{kA}{4}$) and standard deviation approximately $\sqrt{\frac{k}{2}} \cdot \frac{A}{2\sqrt{3}}$.

Thus, for N and k sufficiently large, the probability that the sum of the positive differences in the first block is in $[\frac{kA}{4} - \frac{\sqrt{kA}}{2\sqrt{6}}, \frac{kA}{4} + \frac{\sqrt{kA}}{2\sqrt{6}}]$ is at least $\frac{1}{2}$ (and a similar statement for the negatives). Thus, of the A^k tuples, at least $\frac{1}{4}A^k$ will have the sums of the positive (negative) differences lying in this interval (in the negative of this interval). We call such choices *good*.

Remember, in the arguments leading up to Equation 28, we only specified two items. First, the *absolute values* of the k differences (all distinct); second, that half the positive differences are in the first block, and the *relative* orderings of the positive (negative) differences in the second block is given.

Thus, we have freedom to choose how to intersperse the positives and negatives in the first and second blocks. Consider a good choice of x_k s. We place these differences in the first block of length k as follows. Choose the first positive difference from our good list, and make the first difference positive. Keep assigning (in order) the positive differences from our good list until the running sum of the differences assigned to the first block exceeds A . Then assign the negative differences from our good list until the running sum of differences in the first block is less than $-A$. We then assign

positive differences again until the running sum exceeds A , and so on. We assign half the positive (negative) differences to the first block.

Throughout the process, the largest the running sum can be in absolute value is $\max(2A, 2 \cdot \frac{\sqrt{k}A}{2\sqrt{6}})$. This is because the $\frac{k}{2}$ positive (negative) differences yield sums whose negatives are very close to each other, and each added difference can change the running sum by at most $\pm A$.

We now assign the differences in the second block. We have already chosen the positive and negative differences. There are $(\frac{k}{2})!$ orderings of the positive (negative) differences. We choose these relative orderings, and now choose how to intersperse these. We put down the differences, again making sure the running sum never exceeds in absolute value $\max(2A, 2 \cdot \frac{\sqrt{k}A}{2\sqrt{6}})$.

Let $i_1 = 0$. From Equation 27, we now see that each index is at most $2 \max(2A, 2 \cdot \frac{\sqrt{k}A}{2\sqrt{6}})$. Therefore, each index is in $\left[-\frac{2}{\sqrt{6}} \frac{N}{k^{\alpha-\frac{1}{2}}}, \frac{2}{\sqrt{6}} \frac{N}{k^{\alpha-\frac{1}{2}}}\right]$. Thus, if we shift i_1 so that $i_1 \in \left[\frac{7}{8} \frac{N}{k^{\alpha-\frac{1}{2}}}, \frac{N}{k^{\alpha-\frac{1}{2}}}\right]$, as $\alpha > \frac{1}{2}$ for k large all indices will now be in $\{1, \dots, N\}$. Thus, this is a valid assignment of indices.

We now count the number of valid assignments. We see this is at least

$$\left(\frac{1}{8} \frac{N}{k^{\alpha-\frac{1}{2}}}\right) \cdot \left(\frac{1}{4} A^k - \binom{k}{2} A^{k-1}\right) \cdot (k/2)!^2. \quad (29)$$

To calculate the contribution to the $2k$ -th moment from this pairing, we divide by N^{k+1} . If any of the differences are the same, there is a slight complication; however, as N is large relative to k , we may remove the small number of cases (at most $\binom{k}{2} A^k$) when we have repeat differences among the \tilde{x}_p s and \tilde{x}_n s. By Stirling's Formula, the main term is

$$\frac{1}{N^{k+1}} \frac{1}{32} \frac{N^{k+1}}{k^{k\alpha-\frac{1}{2}}} \left(e^{\frac{k}{2} \log \frac{k}{2} - \frac{k}{2}} \sqrt{2\pi(k/2)}\right)^2 = \frac{\pi k^{\frac{3}{2}}}{16e^{(1+\log 2)k}} \cdot e^{(1-\alpha)k \log k}. \quad (30)$$

Thus, the $2k$ -th root looks like $\frac{e^{(1-\alpha) \log k}}{e^{1+\log 2}} > O(k^{1-\alpha})$, proving the support is unbounded.

5. WEAK CONVERGENCE

We need to show that the variances tend to 0. Thus, we must show

$$\lim_{N \rightarrow \infty} \left(\mathbb{E}[M_m(A, N)^2] - \mathbb{E}[M_m(A, N)]^2 \right) = 0. \quad (31)$$

As $M_m(A, N) = \frac{1}{N^{\frac{m}{2}+1}} \text{Trace}(A^m)$, we have

$$\begin{aligned}
\mathbb{E}[M_m(A, N)^2] &= \frac{1}{N^{m+2}} \sum_{1 \leq i_1, \dots, i_m \leq N} \sum_{1 \leq j_1, \dots, j_m \leq N} \mathbb{E}[b_{|i_1-i_2|} \cdots b_{|i_m-i_1|} b_{|j_1-j_2|} \cdots b_{|j_m-j_1|}] \\
\mathbb{E}[M_m(A, N)]^2 &= \frac{1}{N^{m+2}} \sum_{1 \leq i_1, \dots, i_m \leq N} \mathbb{E}[b_{|i_1-i_2|} \cdots b_{|i_m-i_1|}] \sum_{1 \leq j_1, \dots, j_m \leq N} \mathbb{E}[b_{|j_1-j_2|} \cdots b_{|j_m-j_1|}].
\end{aligned} \tag{32}$$

There are two possibilities: if the absolute values of the differences from the i s are completely disjoint from those of the j s, then these contribute equally to $\mathbb{E}[M_m(A, N)^2]$ and $\mathbb{E}[M_m(A, N)]^2$. We are left with estimating the difference for the crossover cases, when the value of an $i_\alpha - i_{\alpha+1} = \pm(j_\beta - j_{\beta+1})$.

We assume $m = 2k$; a similar proof works for odd m . Note $N^{m+2} = N^{2k+2}$. The following two lemmas imply the variance tends to 0. As our moments grow slower than the Gaussian, we satisfy the conditions necessary to obtain almost surely weak convergence.

Lemma 5.1. *The contribution from crossovers in $\mathbb{E}[M_{2k}(A, N)]^2$ is $O_k(\frac{1}{N})$.*

Proof. For $\mathbb{E}[M_{2k}(A, N)]$, the expected value vanishes if anything is unpaired. Thus, in $\mathbb{E}[M_{2k}(A, N)]^2$, in the i s and j s everything is at least paired, and there is at least one common value from a crossover. The maximum number of such possibilities occurs when everything is paired on each side, and just one set of pairs crosses over; for this crossover there are 2 ways to choose sign. In this case, there are $k + 1$ degrees of freedom in the i s, and $k + 1 - 1$ degrees of freedom in the j s (we lost one degree of freedom from the crossover). Thus, these terms give $O(N^{2k+1})$. Considering now matchings on each side with triple or higher pairings, more crossovers, and the two possible assignments of sign to the crossovers, we find that i s and j s with a crossover contribute $O_k(\frac{1}{N})$ to $\mathbb{E}[M_m(A, N)]^2$. \square

Lemma 5.2. *The contribution from crossovers in $\mathbb{E}[M_m(A, N)]^2$ is $O_k(\frac{1}{N})$.*

Proof. If neither the i differences nor the j differences have anything unpaired (ie, everything is either paired or higher), and there is at least one crossover, it is easy to see these terms are $O_k(\frac{1}{N})$. The difficulty occurs when we have unmatched singletons on either side. Assume there are unmatched differences among the i s. We only increase the number of degrees of freedom by replacing triple pairings and higher among the i s with pairs and singletons (note we may lose these degrees of freedom as these must be crossed and matched with the j s, but we can always cross these over to the j s with no net loss of degrees of freedom). Similarly, we can remove triple and higher pairings among the j s.

Assume there are $s_i > 0$ singletons and $k - \frac{s_i}{2}$ pairs on the i side, $s_j \geq 0$ singletons on the j side, and $C \geq \max(s_i, s_j)$ crossings. Note s_j can equal 0, if we send the singletons on the i side to matched pairs among the j s, but C cannot be less than s_i and s_j . Note s_i, s_j are even.

On the i side, there are $1 + (k - \frac{s_i}{2}) + (s_i - 1)$ degrees of freedom; the 1 is from the freedom of assigning any value to one index, then we have $k - \frac{s_i}{2}$ from pairs, and then the last singleton's value is determined, so we have just $s_i - 1$ additional degrees of freedom from singletons.

Assume $s_j > 0$. On the j side, there could have been $1 + (k - \frac{s_j}{2}) + (s_j - 1)$ degrees of freedom, but we know we have C crossings. This loses at least $C - 1$ degrees of freedom (it's possible the last, forced j difference already equalled an i difference). Thus, the number of degrees of freedom is

$$\left[1 + \left(k - \frac{s_i}{2}\right) + (s_i - 1)\right] + \left[1 + \left(k - \frac{s_j}{2}\right) + (s_j - 1) - (C - 1)\right] = 2k + 1 - \frac{1}{2}(2C - s_i - s_j). \quad (33)$$

If $s_j = 0$, then there are $1 + k - C$ degrees of freedom on the j side, and we get $2k + 1 - (C - \frac{s_i}{2})$ degrees of freedom.

Thus, there are at most $2k + 1$ degrees of freedom. Doing the combinatorics for choices of sign and number of triples and higher shows these terms contribute $O_k(\frac{1}{N})$. \square

Theorem 5.3. *The measures $\mu_{A,N}(x)$ weakly converge to a universal measure of unbounded support, independent of p .*

Proof. As M_k is less than the Gaussian's moments, the M_k s uniquely determine a probability measure, which by Section 4 has unbounded support. As $\mathbb{E}[M_k(A, N)] \rightarrow M_k$ and the variances tend to zero, standard arguments give weak convergence. \square

6. ALMOST SURE CONVERGENCE

6.1. Expansions. For convenience in presentation, we assume $p(x)$ is even (ie, the odd moments vanish); we remark on the trivial modifications to handle the additional book-keeping from general $p(x)$. We will show

$$\lim_{N \rightarrow \infty} \mathbb{E} [|M_m(A, N) - \mathbb{E}[M_m(A, N)]|^4] = O\left(\frac{1}{N^2}\right). \quad (34)$$

The above (plus Chebychev and Borel-Cantelli) will yield almost sure convergence. Expanding this out, it is sufficient to study

$$\begin{aligned} & \mathbb{E}[M_m(A, N)^4] - 4\mathbb{E}[M_m(A, N)^3]\mathbb{E}[M_m(A, N)] + 6\mathbb{E}[M_m(A, N)^2]\mathbb{E}[M_m(A, N)]^2 \\ & - 3\mathbb{E}[M_m(A, N)]\mathbb{E}[M_m(A, N)]^3. \end{aligned} \quad (35)$$

For even moments, we may write the pieces as

$$\begin{aligned}\mathbb{E}[M_{2m}(A, N)^4] &= \frac{1}{N^{4m+4}} \sum_i \sum_j \sum_k \sum_l \mathbb{E}[b_{is}b_{js}b_{ks}b_{ls}] \\ \mathbb{E}[M_{2m}(A, N)^3]\mathbb{E}[M_{2m}(A, N)] &= \frac{1}{N^{4m+4}} \sum_i \sum_j \sum_k \sum_l \mathbb{E}[b_{is}b_{js}b_{ks}]\mathbb{E}[b_{ls}],\end{aligned}\tag{36}$$

(note we combined the $\binom{4}{3}$ and $\binom{4}{4}$ terms) and so on, where for instance

$$E_1 = \mathbb{E}[b_{is}b_{js}b_{ks}b_{ls}] = \mathbb{E}[b_{|i_1-i_2|} \cdots b_{|j_{2m}-j_1|} b_{|k_1-k_2|} \cdots b_{|k_{2m}-k_1|} b_{|l_1-l_2|} \cdots b_{|l_{2m}-l_1|}].\tag{37}$$

We fix some notation. Denote the expected value sums above by E_1, E_2, E_3 and E_4 (which occur with factors of 1, -4 , 6 and -3 respectively). For $h \in \{i, j, k, l\}$, let b_h refer to the differences in $b_{|h_1-h_2|} \cdots b_{|h_{2m}-h_1|}$. If a difference in a b_h is matched with another difference in b_h , we say this is an *internal* matching; otherwise, it is an *external* matching. By a singleton, pair, triple, quadruple and so on, we refer to matchings within a b_h (ie, an internal matching). Thus, a triple occurs when exactly three of the differences in a b_h are equal.

Let p_a denote the a -th moment of $p(x)$. Note $p_2 = 1$. In $\sum \mathbb{E}[b_i b_j b_k b_l]$, if we have all differences occurring twice, except for two different differences occurring four times (two quadruples) and another different one occurring six times (one sextuple), we would have $1^{2m-7} p_4^2 p_6$.

Note there are at most $4m + 4$ degrees of freedom – everything must be matched in at least pairs (we have $8m$ total differences, as we are looking at the fourth power of the $2m$ -th moment), and then each b_h has at most one more degree of freedom (can choose any index). Thus, any terms with a loss of at least two degrees of freedom contribute at most $O(\frac{1}{N^2})$.

6.2. Only Pairs and Singletons. We show there is no net contribution if there are no triples or higher, and then deal with that case afterwards.

Lemma 6.1. *Assume in addition there are no singletons. Then the contribution is $O(\frac{1}{N^2})$.*

Proof. If there are no matchings between b_h s, then everything is independent, and we get $1 - 4 + 6 - 3 = 0$. If two pairs are matched, we lose one degree of freedom. There are $\binom{4}{2} = 6$ ways to choose two out of i, j, k, l to share a match.

For the four expected value sums, we get the following contributions: $\binom{4}{2}p_4$ from E_1 ; $\binom{3}{2}p_4 + (6 - \binom{3}{2})$ from E_2 (three times the two pairs are

in the expected value of a product together, giving p_4 ; the other three times they are separated, giving $p_2 = 1$); $\binom{2}{2}p_4 + (6 - \binom{2}{2})$ from E_3 (only once are the matched pairs together); $\binom{4}{2}$ from E_4 . Combining yields

$$1 \cdot 6p_4 - 4(3p_4 + 3) + 6(p_4 + 5) - 3(6) = 0. \quad (38)$$

If at least three pairs are matched together, or two sets of two pairs are matched together, we lose at least 2 degrees of freedom, giving a contribution of size $O(\frac{1}{N^2})$. \square

The following lemmas are the cornerstone of the later combinatorics:

Lemma 6.2. *If there is a singleton in b_h paired with something in b_g , then there is a loss of at least one degree of freedom.*

Note if every difference in a b_h (all singletons) is paired with a difference in b_g (all singletons), we have a loss of one degree of freedom. We can choose any index and $2m - 1$ differences in b_h ; the last difference is now determined. Once we choose one index in b_g , all other indices are determined, for a total of $1 + (2m - 1) + 1$ (instead of $2m + 2$) degrees of freedom. Thus, instead of being able to choose $2m$ differences freely, we could only choose $2m - 1$.

Note the above argument holds if instead of all singletons, we have elements of b_g and b_h only matched internally and externally with each other.

Proof. As we can cycle the labels, we may assume that $b_{|h_{2m}-h_1|}$ is the singleton. Note that once any index and the values of the other differences are given, then $|h_{2m} - h_1|$ is determined. We would like to conclude it is not free, and we have lost a degree of freedom.

Its value is forced, and it must equal the difference from another b_g ($h \neq g \in \{i, j, k, l\}$), say $b_{|g_a-g_{a+1}|}$. If $b_{|g_a-g_{a+1}|}$ wasn't forced, we have just lost a degree of freedom; if it was forced, then we have already lost a degree of freedom. \square

Remark 6.3. *In the above, we did not need the matching to be with a singleton – a pair, triple or higher would also have worked.*

Lemma 6.4. *If at least three of the b_h s have a singleton, there is a loss of at least two degrees of freedom.*

Proof. If there is a matching of singletons from say b_i and b_j , and another matching from b_k and b_l , the lemma is clear from above. Without loss of generality, the remaining case is when a singleton from b_i is matched with one from b_j , and another singleton from b_i is matched with one from b_l . We then apply the previous lemma to (b_j, b_i) and (b_k, b_i) . \square

We can now prove

Theorem 6.5. *The contribution when there are no triple or higher internal pairings is at most $O(\frac{1}{N^2})$.*

Proof. It is sufficient to show the non-zero contributions all lost at least two degrees of freedom. We have already handled the case when there are no singletons. If three or four b_h s have a singleton, we are done by Lemma 6.4. If exactly two have singletons, then there is no contribution in the E_1 through E_4 , except for the cases when they are under the expected value together (remember the mean of p vanishes).

We have already lost a degree of freedom in this case; if any pair in any b_h is matched with a pair in a b_g , we lose another degree of freedom. Thus, we may assume there are no matches with four or more elements. Thus, every difference that occurs, occurs exactly twice.

There are $\binom{4}{2} = 6$ ways to choose which two of the four b_h s have singletons paired. The contribution from E_1 is 6, from E_2 is 3 (3 of the 6 times they are under the expected value together; the other 3 times they are separated, and the expected value of a difference occurring once is 0), from E_3 is 1 (only 1 of the 6 ways have them under the expected value together), and from E_4 is 0. Thus, we have a contribution of

$$1 \cdot 6 - 4 \cdot 3 + 6 \cdot 1 - 3 \cdot 0 = 0. \quad (39)$$

We are left with the case when the only singletons are in one b_h . As we are assuming there are no triple or higher internal matchings, these singletons must then be matched with pairs, giving external triples; as the odd moments of $p(x)$ vanish, there is no net contribution. \square

Remark 6.6. *If we do not assume the odd moments of p vanish, additional book-keeping yields the contribution is of size $\frac{1}{N^2}$. If exactly two of the b_h s have singletons, then each has at least two; we've already handled the case when they are matched together. As no difference can be left unmatched, we just need to study the case when we get four triples or two triples and a pair; each clearly loses two degrees of freedom;*

We are left with the case when only one b_h has singletons. We are down one degree of freedom already, so there cannot be another non-forced matching. If there are at least four singletons, we are done. If there are two singletons, we get two triples (either with the same or different b_g s). Similar arguments as before yield the contributions are

$$1 \cdot 6p_3^2 - 4 \cdot 3p_3^2 + 6 \cdot p_3^2 - 3 \cdot 0 = 0 \quad (40)$$

if the two external triples involve matchings from b_h to the same b_g , and

$$1 \cdot 4p_3^2 - 4 \cdot 3p_3^2 + 6 \cdot 0 - 3 \cdot 0 = 0. \quad (41)$$

6.3. Eliminating Triple and Higher Matchings.

Lemma 6.7. *If there are no crossovers, there is no net contribution.*

Proof. If there are no crossovers, the expected value of the products are the products of the expected values. Thus, each term becomes $\mathbb{E}[M_{2m}(A, N)]^4$, and $1 - 4 + 6 - 3 = 0$. \square

Lemma 6.8. *If there are at least two triples among all of the b_h s, the contribution is $O\left(\frac{1}{N^2}\right)$.*

Proof. Everything must be matched in at least pairs (or its expected value vanishes). If there are only two values among six differences, then instead of getting 3 degrees of freedom, we get 2. This is enough to see decay like $O\left(\frac{1}{N}\right)$. If we didn't assume $p(x)$ were even, we would have more work; as the odd moments vanish, however, the two triples must be paired with other differences, or with each other. In either case, we lose at least one degree of freedom from each, completing the proof. \square

Remark 6.9. *Similarly, one can show there cannot be a triple and anything higher than a triple. Further, we cannot have two quadruples or more, as a quadruple or more loses one degree of freedom (a quadruple is two pairs that are equal – instead of having two degrees of freedom, we now have one).*

Lemma 6.10. *If there is a quadruple, quintuple, or higher matchings within a b_h , the contribution is $O\left(\frac{1}{N^2}\right)$.*

Proof. There can be no sextuple or higher, as this gives at least three pairs matched, yielding one degree of freedom (instead of three). If there is a quadruple or quintuple, everything else must be pairs or singletons. As the odd moments vanish, a quintuple must be matched with at least a singleton, again giving six points matched, but only one degree of freedom.

We are left with one quadruple (which gives a loss of one degree of freedom) and all else pairs and singletons. No pairs can be matched to the quadruple or each other, as we would then lose at least two degrees of freedom. If there are any singletons, by Lemma 6.2 there is a loss of a degree of freedom. If we have a quintuple or higher, this is enough to lose two degrees of freedom. Thus, we need only study the case of all pairs and one quadruple, with no external matchings.

As everything is independent, we find a contribution of

$$1 \cdot p_4 - 4 \cdot p_4 + 6 \cdot p_4 - 3p_4 = 0, \quad (42)$$

where p_4 is the fourth moment of p . \square

Lemma 6.11. *If there is only one triple (say in b_h), the contribution is $O(\frac{1}{N^2})$.*

Proof. As odd moments vanish, the triple must be paired with a singleton from another b_h ; further, there must be at least one singleton in the same b_h as the triple (as there are an even number of terms). We thus lose a degree of freedom from the triple matched with a singleton (four points, but one instead of two matches), and we lose a degree of freedom from the singleton in the same b_h as the triple (Lemma 6.2). Thus, we have lost two degrees of freedom. \square

We have proved

Theorem 6.12. *The contribution from having a triple or higher internal matching is $O(\frac{1}{N^2})$.*

Remark 6.13. *Similar arguments work for general $p(x)$.*

6.4. Odd Moments. As the odd moments of $p(x)$ vanish, handling

$$\lim_{N \rightarrow \infty} \mathbb{E} [|M_{2m+1}(A, N) - \mathbb{E}[M_{2m+1}(A, N)]|^4] = O\left(\frac{1}{N^2}\right) \quad (43)$$

is significantly easier.

Theorem 6.14. *We lose at least two degrees of freedom above, implying the expected value is $O(\frac{1}{N^2})$.*

Proof. In each b_h , there is at least one odd internal matching (or singleton); thus, only E_1 can be non-zero. If there are four (or more) internal triples (or higher), we lose at least two degrees of freedom.

If there are exactly three internal triples, either two are matched together and one is matched with a singleton, or all three are matched with singletons; in both cases we lose at least two degrees.

If there are exactly two internal triples, there must be at least two b_h s with singletons. If the triples are matched with singletons, we lose two degrees; if the triples are matched together we lose one degree from that, and one more degree from the singletons (Lemma 6.2).

If there is exactly one triple, at least three b_h s have singletons, and similar arguments yield a loss of at least two degrees.

If there are no triples, then by Lemma 6.4 there is a loss of at least two degrees. \square

Combining Theorems 6.12 and 6.14 yields

Theorem 6.15.

$$\lim_{N \rightarrow \infty} \mathbb{E} [|M_m(A, N) - \mathbb{E}[M_{2m+1}(A, N)]|^4] = O\left(\frac{1}{N^2}\right). \quad (44)$$

Remark 6.16. *Similar arguments work for general $p(x)$.*

6.5. Almost Sure Convergence. We show that we have almost sure convergence. We first introduce some notation, and then show how this follows from Theorem 6.15.

Fix $p(x)$ as before. Let Ω_N be the outcome space $(T_N, \prod_{i=1}^{N-1} p(b_i)db_i)$, where T_N is the space of all $N \times N$ Real Symmetric Toeplitz matrices. Let Ω be the outcome space $(T_{\mathbb{N}}, \prod p)$, where $T_{\mathbb{N}}$ is the set of all $\mathbb{N} \times \mathbb{N}$ Real Symmetric Toeplitz matrices and $\prod p$ is the product measure built from having the entries iidrv from $p(x)$. For each N , we have projection maps from Ω to Ω_N . Thus, if $A \in T_{\mathbb{N}}$ is a Real Symmetric Toeplitz matrices, then A_N is the restriction obtained by looking at the upper left $N \times N$ block of A .

We slightly adjust some notation from before. Let $\mu_{A_N}(x)dx$ be the probability measure associated to the Toeplitz $N \times N$ matrix A_N . Then

$$\begin{aligned} \mu_{A_N}(x)dx &= \frac{1}{N} \sum_{i=1}^N \delta \left(x - \frac{\lambda_i(A_N)}{\sqrt{N}} \right) \\ M_m(A_N) &= \int_{\mathbb{R}} x^m \mu_{A_N}(x)dx \\ M_m(N) &= \mathbb{E}[M_m(A_N)] \\ M_m &= \lim_{N \rightarrow \infty} M_m(N). \end{aligned} \tag{45}$$

As $N \rightarrow \infty$, $M_m(N)$ converges to M_m , and the convergence for each m is at the rate of $\frac{1}{N}$. The expectation above is with respect to the product measure on T_N built from $p(x)$.

We want to show that, for all m , as $N \rightarrow \infty$,

$$M_m(A_N) \longrightarrow M_m \text{ almost surely.} \tag{46}$$

By the triangle inequality,

$$|M_m(A_N) - M_m| \leq |M_m(A_N) - M_m(N)| + |M_m(N) - M_m|. \tag{47}$$

As the second term tends to zero, it is sufficient to show the first tends to zero for almost all A .

Chebychev's Inequality states that for any random variable X with mean zero and finite m -th moment that

$$\text{Prob}(|X| \geq \epsilon) \leq \frac{\mathbb{E}[X^m]}{\epsilon^m}. \tag{48}$$

Note $\mathbb{E}[M_m(A_N) - M_m(N)] = 0$, and by Theorem 6.15, $M_m(A_N) - M_m(N)$ has finite fourth moment. In fact, Chebychev's Inequality and Theorem 6.15 yield

$$\text{Prob}(|M_m(A_N) - M_m(N)| \geq \epsilon) \leq \frac{\mathbb{E}[|M_m(A_N) - M_m(N)|^4]}{\epsilon^4} \leq \frac{C_m}{N^2 \epsilon^4}. \quad (49)$$

The proof is completed by applying the following:

Lemma 6.17 (Borel-Cantelli). *Let B_i be a sequence of events with $\sum_i \text{Prob}(B_i) < \infty$. Let*

$$B = \left\{ \omega : \omega \in \bigcap_{j=1}^{\infty} \bigcup_{k=j}^{\infty} B_i \right\}. \quad (50)$$

Then the probability of B is zero.

In other words, an ω is in B if and only if that ω is in infinitely many B_i , and the probability of events ω which occur infinitely often is zero.

Fix a large k and let

$$B_N^{(k,m)} = \{A \in T_N : |M_m(A_N) - M_m(N)| \geq \frac{1}{k}\}. \quad (51)$$

We have seen that $\text{Prob}(B_N^{(k,m)}) \leq \frac{C_m k^4}{N^2}$. Thus, for fixed m and k , the conditions of the Borel-Cantelli Lemma are met, and we deduce that the probability of $A \in T_N$ that occur in infinitely many $B_N^{(k,m)}$ is zero. We now let $k \rightarrow \infty$, and find for any fixed m , as $N \rightarrow \infty$, $M_m(A_N) \rightarrow M_m$ with probability one. Let $B_m^{i.o.}$ be the probability zero sets where we do not have such convergence.

Let $B^{i.o.} = \bigcup_{m=1}^{\infty} B_m^{i.o.}$. As a countable union of probability zero sets has probability zero, we see that $\text{Prob}(B^{i.o.}) = 0$; however, this is precisely the set where for some m , we do not have pointwise convergence.

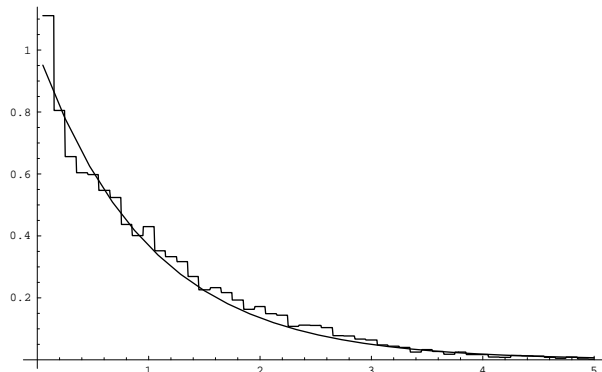
Thus, except for a set of probability zero, we find $M_m(A_N) \rightarrow M_m$ for all m .

7. POISSONIAN BEHAVIOR?

As there are only $N - 1$ degrees of freedom for the Toeplitz Ensemble, and not $O(N^2)$, it is reasonable to believe the spacings between adjacent normalized eigenvalues may differ from those of full Real Symmetric Matrices. For example, band matrices of width 1 are just diagonal matrices, and there the spacing is Poissonian (e^{-x}); full Real Symmetric Matrices are conjectured to have their spacing given by the GOE distribution (which is well approximated by Axe^{-Bx^2}).

For d -regular graphs, there are $\frac{dN}{2}$ degrees of freedom. It has been numerically observed (see [JMRR] among others) that the spacings between adjacent eigenvalues look GOE.

We chose 1000 Toeplitz matrices (1000×1000), with entries iidrv from the standard normal. We looked at the spacings between the middle 11 normalized eigenvalues for each matrix, giving us 10 spacings. A plot of the spacings between normalized eigenvalues looks Poissonian.



We conjecture that in the limit as $N \rightarrow \infty$, the local spacings between adjacent normalized eigenvalues will be Poissonian. It is interesting to note that Random d -Regular Graphs have a comparable number of degrees of freedom; however, in their adjacency matrices, there is significantly more independence in the a_{ij} – for the Toeplitz Ensemble, we have a strict structure, namely a_{ij} depends only on $|i - j|$.

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